

# Optimization of online direct marketing efforts

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# Definitions

- Lucid
  - Easily understood; intelligible.
  - Mentally sound; sane or rational.
  - Translucent or transparent.
- Limpid
  - Characterized by transparent clearness; pellucid.
  - Easily intelligible; clear: *writes in a limpid style.*
  - Calm and untroubled; serene.



# Test 1: Two Email campaigns

- Target:
  - Email 1: Males who had registered to sweepstakes 1.
  - Email 2: Males and Females who had registered to the sweepstakes 2.
- Content:
  - Email 1: Sex and Mayhem
  - Email 2: Power and relationship



# Raw Results

	Impression		Click   Impression	
	Email 1	Email 2	Email 1	Email 2
Male	42%	41%	20%	14%
Female		28%		18%



## Test 2

- Target:
  - Registrants from other Video Releases.
  - Registrants from the Movie website.
- Content:
  - Video 1
  - Video 2



## Test 2: Raw Results

	Video 1	Video 2
Men	20%*	14%
Women	11%	18%

\* Click-through rates on targeted emails



## Traditional DM Testing

- Guess a response rate
- Set a confidence level
- Set an acceptable margin of error
- Use *probability tables* to set test size
- Send test
- Wait, wait wait
- Use *probability tables* to evaluate test



## Online DM testing

- Do the same
- Do better
  - Constrained testing
  - Dynamic testing

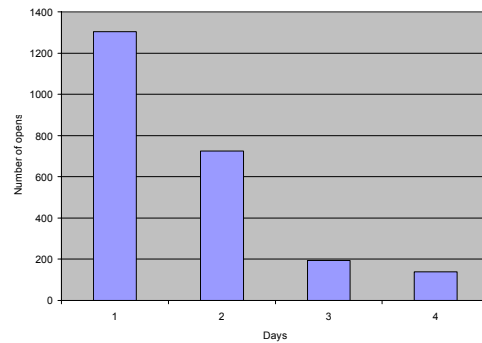


## Constrained testing

- Weekly newsletters, limited time for:
  - Writing
  - Testing
  - Sending



## Open time distribution



## A model of constrained testing

- M emails to send
- T periods to send the whole campaign
- r emails per hour
- 2 possible emails with click probabilities of p1, p2
- How many emails should be used for testing purposes, how long should the test last?



## How long should the test last?

- If we use N emails for testing purposes then we can use  $T_1$  periods for testing:

$$T_1 = T - \frac{M - N}{r}$$



## Test size?

- How many opens (S) can we expect given N and  $T_1$ ?
- What is the power of a test with size S?
- How do we balance return from testing and production?



## Expected opens: S

- Simple case:
  - Send is instantaneous
  - Open time is  $\text{Expo}(\lambda)$

$$f(x) = \lambda e^{-\lambda x}$$

$$F(x) = 1 - e^{-\lambda x}$$

$$\Rightarrow S = E[\text{open} | N, x] = N(1 - e^{-\lambda x})$$



## Send is not instantaneous

- Send rate is r, for the  $i^{\text{th}}$  email we have:

$$f(x, r, i) = \lambda e^{-\lambda(x - \frac{i}{r})} I_{(x, i/r)}$$

$$I_{(x, i/r)} = 1, \forall x \geq i/r$$

$$I_{(x, i/r)} = 0, \forall x < i/r$$

$$F(x, r, i) = 1 - e^{-\lambda(x - \frac{i}{r})}, \text{ if } x > i/r$$



## Expected open: S

$$\begin{aligned} S &= E[\text{open} | N, x, r] = \sum_{i=1}^N F(x, r, i) \\ &= \sum_{i=1}^N 1 - e^{-\lambda(x - i/r)} \\ &= \frac{e^{\lambda/r} - e^{\lambda/r + \lambda N/r} - N e^{\lambda x} + N e^{\lambda/r + \lambda x}}{e^{\lambda x} (e^{\lambda/r} - 1)} \\ &= N - \frac{(e^{\frac{\lambda N}{r}} - 1)}{e^{\lambda(x - \frac{1}{r})} (e^{\frac{\lambda}{r}} - 1)} \end{aligned}$$



## Power of test of size S

$$\alpha = \Phi \left( \frac{P_1 - P_2}{\sqrt{P(1-P) \frac{4}{S}}} \right)$$

- where P is the average of  $p_1, p_2$  and  $\Phi$  is the CDF of the standard normal distribution.



## Expected revenue for the test

$$ER(M, N, T, r, \lambda, p_1, p_2) = \underbrace{NP}_{\text{Testing}} + \underbrace{(M-N)(\alpha p_1 + (1-\alpha)p_2)}_{\text{Production}}$$

- We can then integrate over the distribution of  $p_1$  and  $p_2$ :

$$ER(M, N, T, r, \lambda) = \int \int [(M-N)(\alpha p_1 + (1-\alpha)p_2) + NP] P(p_1 = P_1) P(p_2 = P_2) dp_1 dp_2$$



## Optimal Size: $N^*$

- All we need to do then is to find  $N^*$  such that:

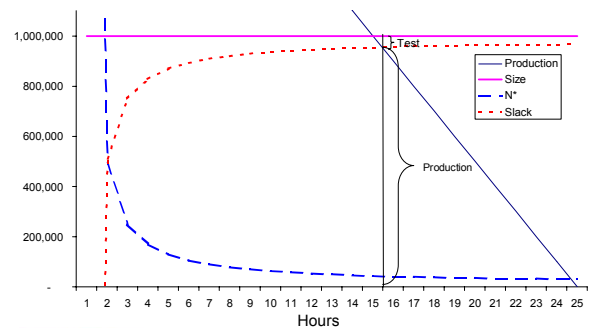
$$\text{Max}_N ER(M, N, T, r, \lambda)$$

$$\text{s.t.} : M \geq N \geq 0$$

$$M \leq Tr$$



## Optimal test: Numerical Solution



## Dynamic testing

- Now that we have a solution to the constrained testing situation, we can implement a dynamic one.
- As the sending occurs, one can update priors on  $p_1$ ,  $p_2$ , and  $\lambda$ .
- As  $p_1$ ,  $p_2$ , and  $\lambda$  are updated, one can update  $N$  and  $T_1$ .



## Implementation issues (1)

- We have ignored the time it takes for people to read the email and click on the links.
- $T_o \gg T_c$
- When solving the constrained problem, we can ignore  $T_c$  because  $T$  is in hours and  $T_c$  in tens of seconds, but if we update as we go, seconds matter!



## Click Time: $T_c$

- $T_c$  is Expo( $\gamma$ )
- Ignore send time
- We can model the time to open and click as BOXMOD:

$$E[\text{Clicks at } t \mid \lambda, \gamma] = N \cdot \frac{1}{\lambda - \gamma} [\lambda - \gamma + \gamma e^{-\lambda t} - \lambda e^{-\gamma t}]$$



## Expected numbers of click

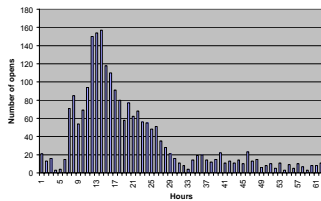
- P(open)
- P(click)

$$E[\text{Click} \mid \lambda, \gamma] = N \cdot \frac{1}{\lambda - \gamma} [\lambda - \gamma + \gamma e^{-\lambda t} - \lambda e^{-\gamma t}] \cdot P(\text{Open}) \cdot P(\text{Click})$$



## Implementation issues (2)

- The open times and click times look exponential from afar, but not from up close:

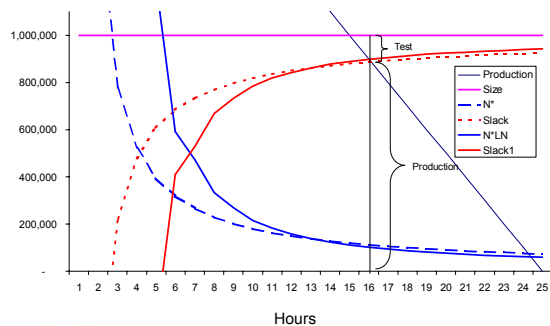


## Time to open/click is Log-Normal

- In the short run, log-normal and expo are very different.
- There is no closed-form expression of the CDF for the log-normal.
- We have to solve numerically or maybe use a log-logistic approximation.



## Log-Normal Solution



## Still to do

- Solve case for log-logistic
- Solve updating formulas for  $P(\text{open})$ ,  $P(\text{Click})$ ,  $\lambda$ , and  $\gamma$
- Simulate test using past data
- Simulate live test
- Live test



## Enhancement

- Stopping rule for failed tests
- ...

